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# Solving Inverse Scattering Problems Using Truncated Cosine Fourier Series Expansion Method

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## 1. Introduction

The aim of inverse scattering problems is to extract the unknown parameters of a medium from measured back scattered fields of an incident wave illuminating the target. The unknowns to be extracted could be any parameter affecting the propagation of waves in the medium.

Inverse scattering has found vast applications in different branches of science such as medical tomography, non-destructive testing, object detection, geophysics, and optics (Semnani & Kamyab, 2008; Cakoni & Colton, 2004).

From a mathematical point of view, inverse problems are intrinsically ill-posed and nonlinear (Colton & Paivarinta, 1992; Isakov, 1993). Generally speaking, the ill-posedness is due to the limited amount of information that can be collected. In fact, the amount of independent data achievable from the measurements of the scattered fields in some observation points is essentially limited. Hence, only a finite number of parameters can be accurately retrieved. Other reasons such as noisy data, unreachable observation data, and inexact measurement methods increase the ill-posedness of such problems. To stabilize the inverse problems against ill-posedness, usually various kinds of regularizations are used which are based on a priori information about desired parameters. (Tikhonov & Arsenin, 1977; Caorsi, et al., 1995). On the other hand, due to the multiple scattering phenomena, the inverse-scattering problem is nonlinear in nature. Therefore, when multiple scattering effects are not negligible, the use of nonlinear methodologies is mandatory.

Recently, inverse scattering problems are usually considered in global optimization-based procedures (Semnani & Kamyab, 2009; Rekanos, 2008). The unknown parameters of each cell of the medium grid would be directly considered as the optimization parameters and several types of regularizations are used to overcome the ill-posedness. All of these regularization terms commonly use a priori information to confine the range of mathematically possible solutions to a physically acceptable one. We will refer to this strategy as the direct method in this chapter.

Unfortunately, the conventional optimization-based methods suffer from two main drawbacks. The first is the huge number of the unknowns especially in 2-D and 3-D cases

which increases not only the amount of computations, but also the degree of ill-posedness. Another disadvantage is the determination of regularization factor which is not straightforward at all. Therefore, proposing an algorithm which reduces the amount of computations along with the sensitivity of the problems to the regularization term and initial guess of the optimization routine would be quite desirable.

2. Truncated cosine Fourier series expansion method

Instead of direct optimization of the unknowns, it is possible to expand them in terms of a complete set of orthogonal basis functions and optimize the coefficients of this expansion in a global optimization routine. In a general 3-D structure, for example the relative permittivity could be expressed as

$$\varepsilon_r(x,y,z)=\sum_{n=0}^{N-1}d_nf_n(x,y,z)$$

(1)

where  $f_n$  is the  $n_{th}$  term of the complete orthogonal basis functions. It is clear that in order to expand any profile into this set, the basis functions must be complete. On the other hand, orthogonality is favourable because with this condition, a finite series will always represent the object with the best possible accuracy and coefficients will remain unchanged while increasing the number of expansion terms. Because of the straightforward relation to the measured data and its simple boundary conditions, using harmonic functions over other orthogonal sets of basis functions is preferable. On the other hand, cosine basis functions have simpler mean value relation in comparison with sine basis functions which is an important condition in our algorithm. We consider the permittivity and conductivity profiles reconstruction of lossy and inhomogeneous 1-D and 2-D media as shown in Fig. 1.

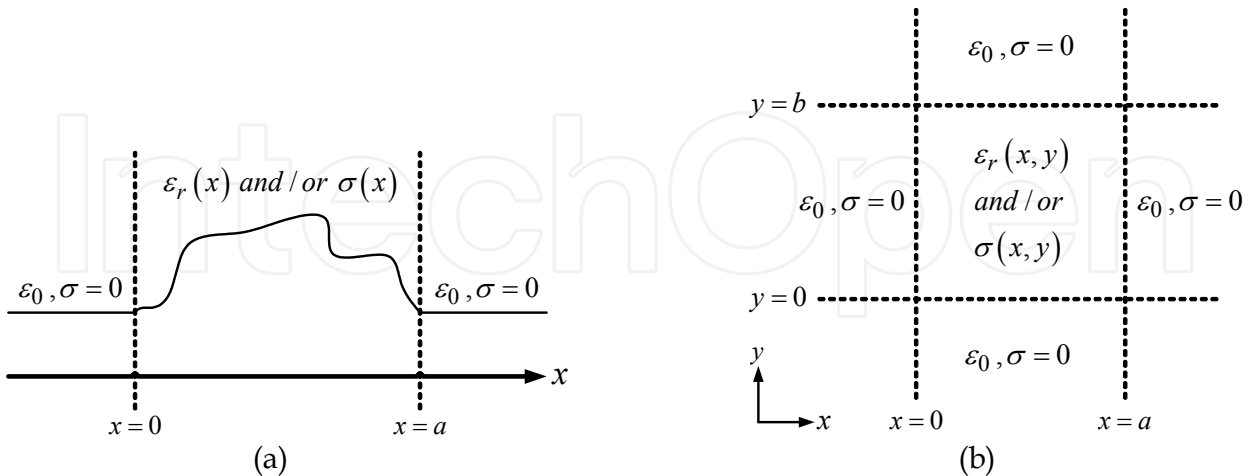


Fig. 1. General form of the problem, (a) 1-D case, (b) 2-D case

If cosine basis functions are used in one-dimensional cases, the truncated expansion of the permittivity profile along  $x$  which is homogeneous along the transverse plane could be expressed as

$$\varepsilon_r(x) = \sum_{n=0}^{N-1} d_n \cos\left(\frac{n\pi}{a}x\right) \tag{2}$$

where  $a$  is the dimension of the problem in the  $x$  direction and the coefficients,  $d_n$ , are to be optimized. In this case, the number of optimization parameters is  $N$  in comparison with conventional methods in which this number is equal to the number of discretized grid points. This results in a considerable reduction in the amount of computations. As another very important advantages of the expansion method, no additional regularization term is needed, because the smoothness of the cosine functions and the limited number of expansion terms are considered adequate to suppress the ill-posedness. In a similar manner for 2-D cases, the expansion of the relative permittivity profile in transverse  $x$ - $y$  plane which is homogeneous along  $z$  can be written as

$$\varepsilon_r(x,y) = \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} d_{nm} \cos\left(\frac{n\pi}{a}x\right) \cos\left(\frac{m\pi}{b}y\right) \tag{3}$$

where  $a$  and  $b$  are the dimensions of the problem in the  $x$  and  $y$  directions, respectively. Similar expansions could be considered for conductivity profiles in lossy cases. The proposed expansion algorithm is shown in Fig. 2. According to this figure, based on an initial guess for a set of expansion coefficients, the permittivity and conductivity are calculated according to the expansion relations like (2) or (3). Then, an EM solver computes a trial electric and magnetic simulation fields. Afterwards, cost function which indicates the difference between the trial simulated and reference measured fields is calculated. In the next step, global optimizer is used to minimize this cost function by changing the permittivity and conductivity of each cell until the procedure leads to an acceptable predefined error.

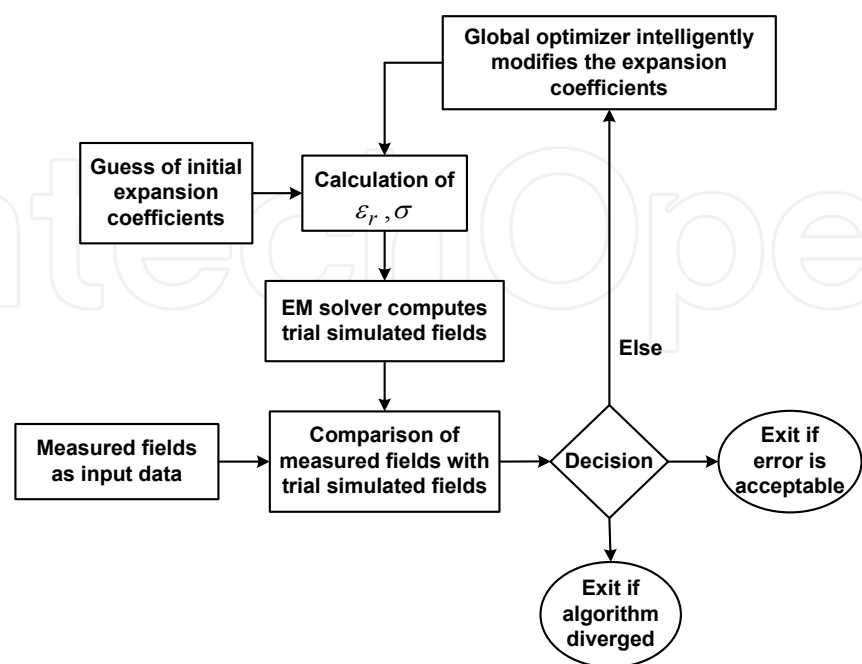


Fig. 2. Proposed algorithm for reconstruction by expansion method

### 3. Mathematical Considerations

As mentioned before, inverse problems are intrinsically ill-posed. Therefore, a priori information must be applied for stabilizing the algorithm as much as possible which is quite straightforward in direct optimization method. In this case, all the information can be applied directly to the medium parameters which are as the same as the optimization parameters. In the expansion algorithm, however, the optimization parameters are the Fourier series expansion coefficients and a priori information could not be considered directly. Hence, a useful indirect routine is vital to overcome this difficulty.

There are two main assumptions about the parameters of an unknown medium. For example, we may assume first that the relative permittivity and conductivity have limited ranges of variation, i.e.

$$1 \leq \varepsilon_r \leq \varepsilon_{r,\max} \quad (4)$$

and

$$0 \leq \sigma \leq \sigma_{\max} \quad (5)$$

The second assumption is that the permittivity and conductivity profiles may not have severe fluctuations or oscillations. These two important conditions must be transformed in such a way to be applicable on the expansion coefficients in the initial guess and during the optimization process.

It is known that average of a function with known limited range is located within that limit, that is if

$$L_1 \leq g(x) \leq L_2, \quad a \leq x \leq b \quad (6)$$

Then

$$L_1 \leq \frac{1}{b-a} \int_a^b g(x) dx \leq L_2 \quad (7)$$

Thus, for 1-D permittivity profile expansion we have

$$1 \leq d_0 \leq \varepsilon_{r,\max} \quad (8)$$

For  $x = 0$ , (2) reduces to

$$\varepsilon_r(0) = \sum_{n=0}^{N-1} d_n \Rightarrow 1 \leq \sum_{n=0}^{N-1} d_n \leq \varepsilon_{r,\max} \quad (9)$$

and for  $x = a$ , we have

$$\varepsilon_r(a) = \sum_{n=0}^{N-1} (-1)^n d_n \Rightarrow 1 \leq \sum_{n=0}^{N-1} (-1)^n d_n \leq \varepsilon_{r,\max} \quad (10)$$

Using Parseval theorem, another relation between expansion coefficients and upper bound of permittivity may be written. For a periodic function  $g(x)$  with period  $T$ , we have

$$\frac{1}{T} \int_T |g(x)|^2 dx = \sum_{n=0}^{\infty} |d_n|^2 \quad (11)$$

Based on (2), (11) may be simplified to

$$1 \leq \sum_{n=0}^{N-1} |d_n|^2 \leq \varepsilon_{r,\max}^2 \quad (12)$$

It is possible to achieve the similar relations for 2-D cases.

$$1 \leq d_{00} \leq \varepsilon_{r,\max} \quad (13)$$

$$1 \leq \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} d_{nm} \leq \varepsilon_{r,\max} \quad (14)$$

$$1 \leq \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} (-1)^{n+m} d_{nm} \leq \varepsilon_{r,\max} \quad (15)$$

$$1 \leq \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} |d_{nm}|^2 \leq \varepsilon_{r,\max}^2 \quad (16)$$

By using the above supplementary equations in the initial guess of the expansion coefficients and as a boundary condition (Robinson & Rahmat-Samii, 2004) during the optimization, the routine converges in a considerable faster rate. Similar conditions can be used for conductivity profiles in lossy cases.

#### 4. Numerical Results

Proposed method stated above is utilized for reconstruction of some different 1-D and 2-D media. In each case, reconstruction by the proposed expansion method is compared with different number of expansion functions in terms of the amount of computations and reconstruction precision.

The objective of the proposed reconstruction procedure is the estimate of the unknowns by minimizing the cost function

$$C = \frac{\sum_{i=1}^I \sum_{j=1}^J \sum_{t=1}^T \left| E_{ij}^{meas}(t) - E_{ij}^{sim}(t) \right|^2}{\sum_{i=1}^I \sum_{j=1}^J \sum_{t=1}^T (E_{ij}^{meas}(t))^2} \tag{17}$$

where  $\vec{E}^{sim}$  is the simulated field in each optimization iteration.  $\vec{E}^{meas}$  is measured field, I and J are the number of transmitters and receivers, respectively and T is the total time of measurement.

To quantify the reconstruction accuracy, the reconstruction errors for example for relative permittivity in 1-D case is defined as

$$e(\varepsilon) = \sqrt{\frac{\sum_{i=1}^{M_x} |\varepsilon_{ri} - \varepsilon_{ri}^o|^2}{\sum_{i=1}^{M_x} (\varepsilon_{ri}^o)^2}} \times 100 \tag{18}$$

where  $M_x$  is the number of subdivisions along x axis and “ $^o$ ” denotes the original scatterer properties.

In all reconstructions in this chapter, FDTD (Taflove & Hagness, 2005) and DE (Storn & Price, 1997) are used as forward EM solver and global optimizer, respectively.

4.1 One-dimensional case

Reconstruction of two 1-D cases is considered in this section. The first one is inhomogeneous and lossless and the second one is considered to be lossy. In the simulations of both cases, one transmitter and two receivers are used around the medium as shown in Fig. 3.

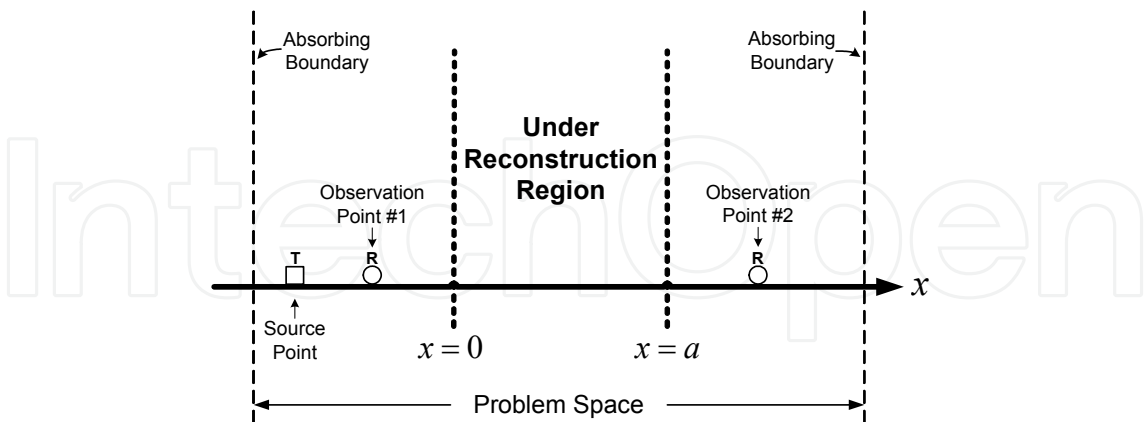


Fig. 3. Geometrical configuration of the 1-D problem

**Test case #1:** In the first sample case, we consider an inhomogeneous and lossless medium consisting 50 cells. Therefore, only the permittivity profile reconstruction is considered. In the expansion method, the number of expansion terms is set to 4, 5, 6 and 7 which results in a lot of reduction in the number of the unknowns in comparison with the direct method. The population in DE algorithm is chosen equal to 100 and the maximum iteration of

optimization is considered to be 300. It must be noted that the initial populations in all reconstruction problems in this chapter are chosen completely random in the solution space. The exact profile and reconstructed ones by the expansion method with different number of expansion terms are shown in Fig. 4a. The variations of cost function (17) and reconstruction error (18) versus the iteration number are plotted in Figs. 4b and 4c, respectively.

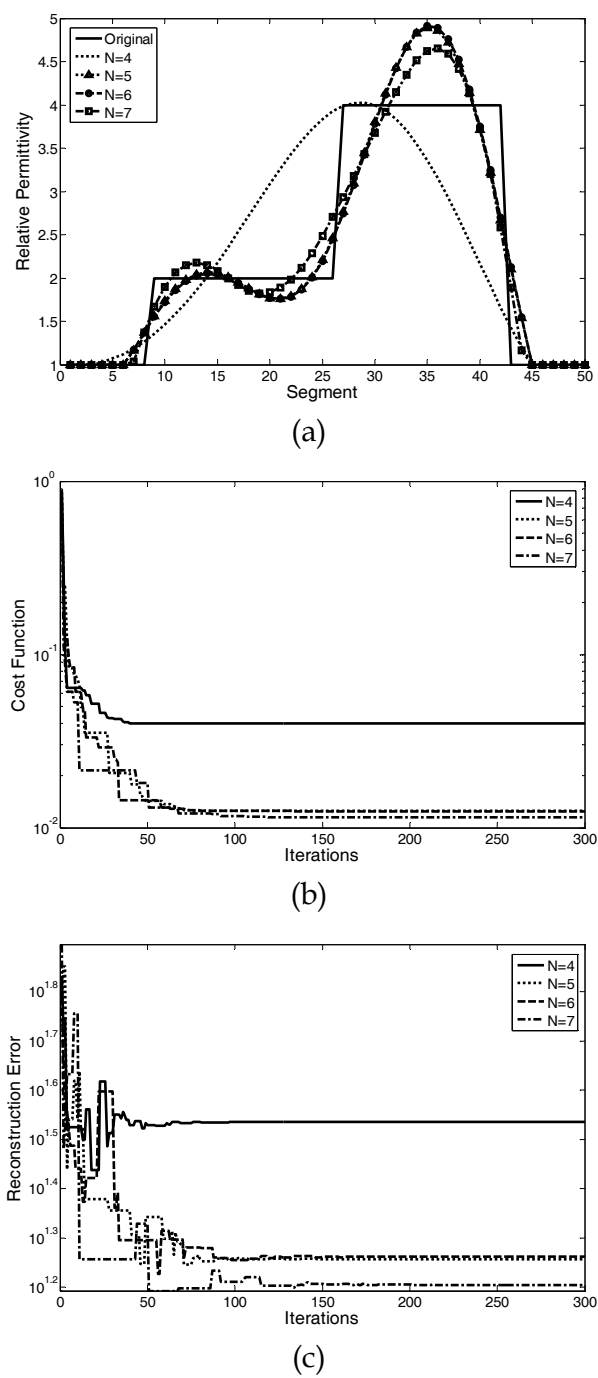
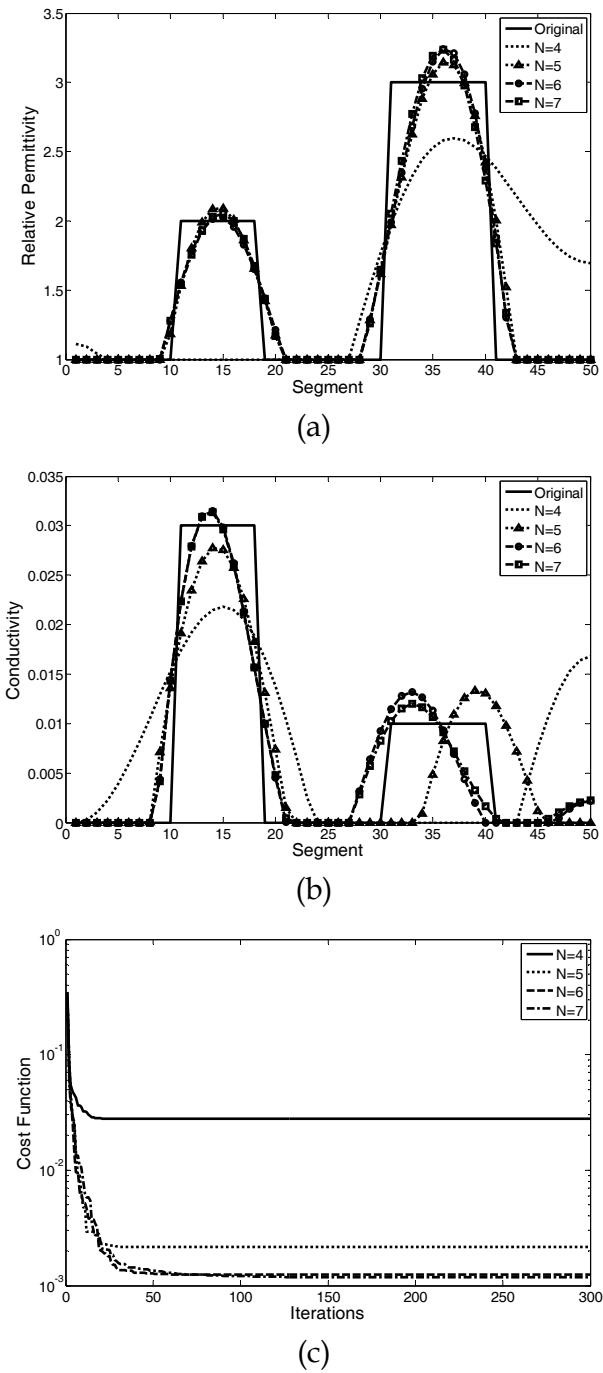


Fig. 4. Reconstruction of 1-D test case #1, (a) original and reconstructed profiles, (b) the cost function and (c) the reconstruction error



**Test case #2:** In this case, a lossy and inhomogeneous medium again with 50 cell length is considered. So, the number of unknowns in direct optimization method is equal to 100. In the expansion method for both permittivity and conductivity profiles expansion,  $N$  is chosen equal to 4, 5, 6 and 7. The optimization parameters are considered equal to the first sample case. The original and reconstructed profiles in addition of the variations of cost function and reconstruction error are presented in Fig. 5.



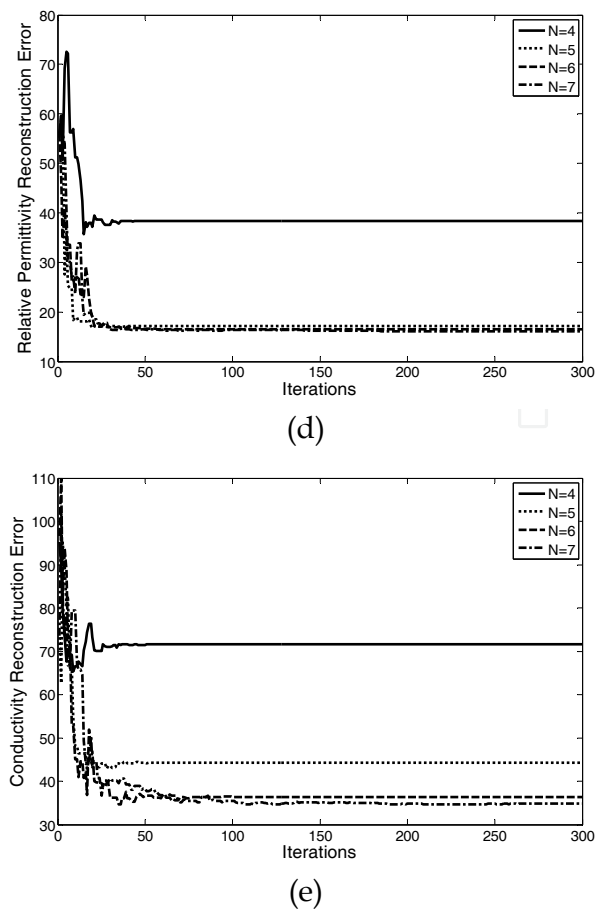


Fig. 5. Reconstruction of 1-D test case #2, (a) original and reconstructed permittivity profiles, (b) original and reconstructed conductivity profiles, (c) the cost function, (d) the permittivity reconstruction error and (e) the conductivity reconstruction error

4.2 Two-dimensional case

The proposed expansion method is also utilized for two 2-D cases. In the simulations of both cases, four transmitter and eight receivers are used as shown in Fig. 6. The population in DE algorithm is chosen equal to 100, the maximum iteration is considered to be 300.

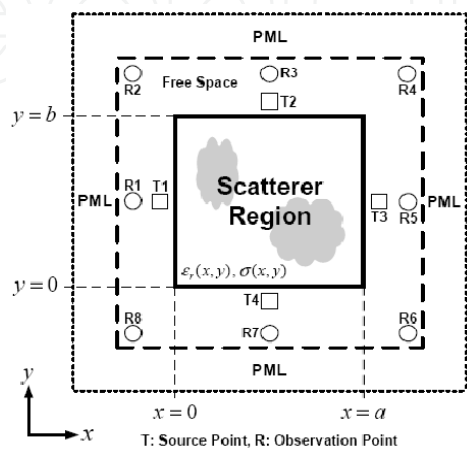


Fig. 6. Geometrical configuration of the 2-D problem

**Case study #1:** In the first sample case, we consider an inhomogeneous and lossless 2-D medium consisting 20\*20 cells. Therefore, only the permittivity profile reconstruction is considered. In the expansion method, the number of expansion terms in both x and y directions are set to 4, 5, 6 and 7. The original profile and reconstructed ones with the use of expansion method are shown in Fig. 7.

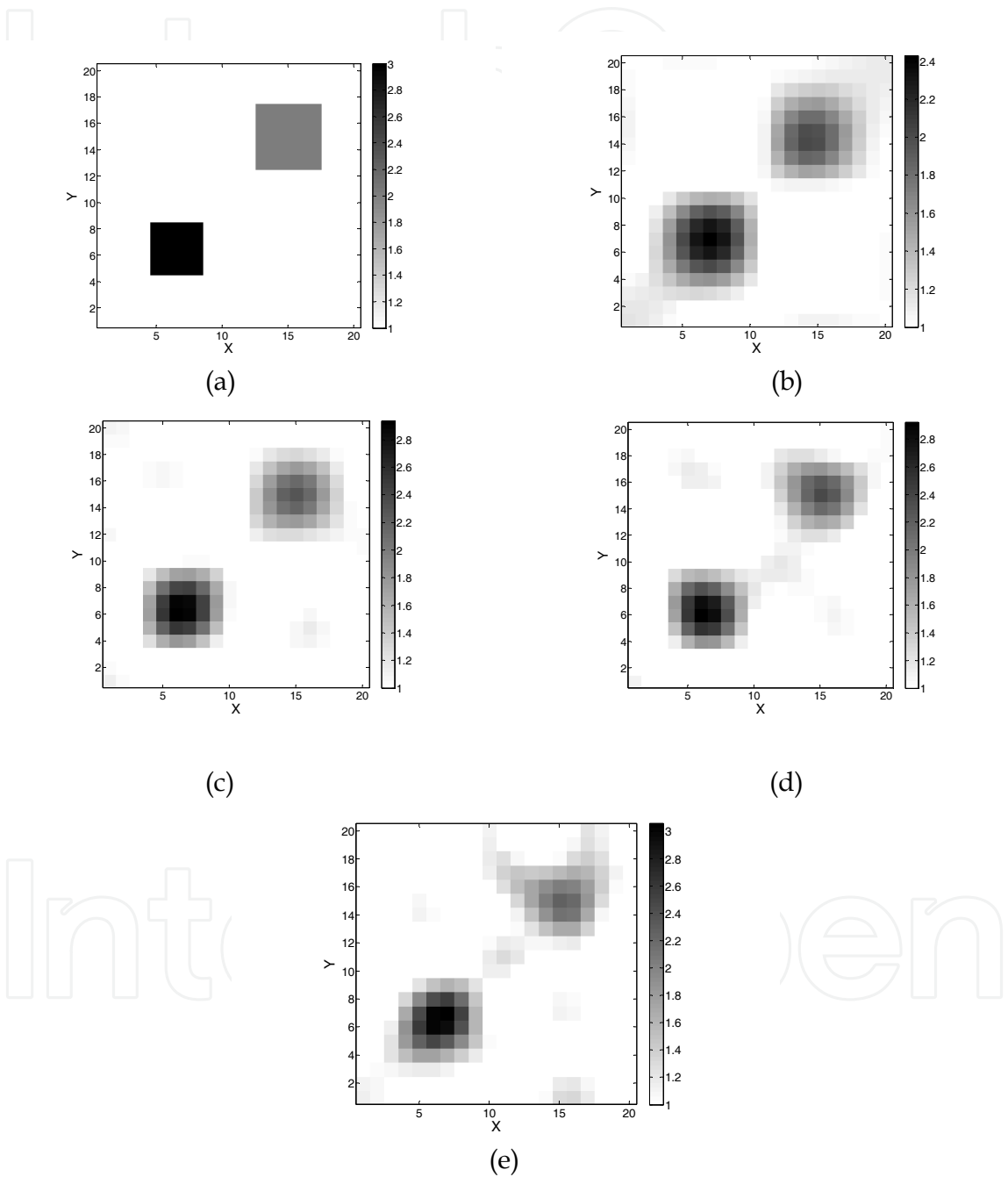


Fig. 7. Reconstruction of 2-D test case #1, (a) original profile, reconstructed profile with (b)  $N=M=4$ , (c)  $N=M=5$ , (d)  $N=M=6$  and (e)  $N=M=7$

The variations of cost function and reconstruction error versus the iteration number are graphed in Fig. 8.

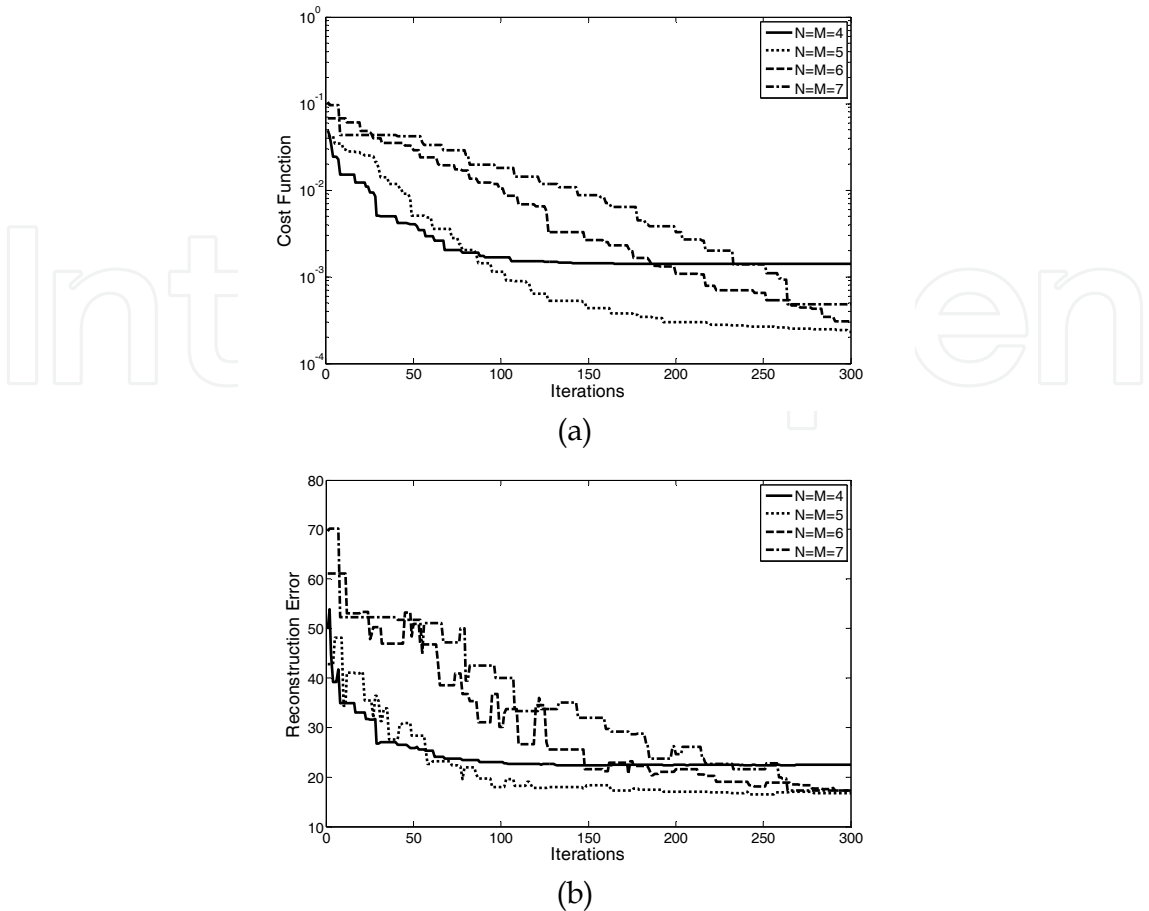
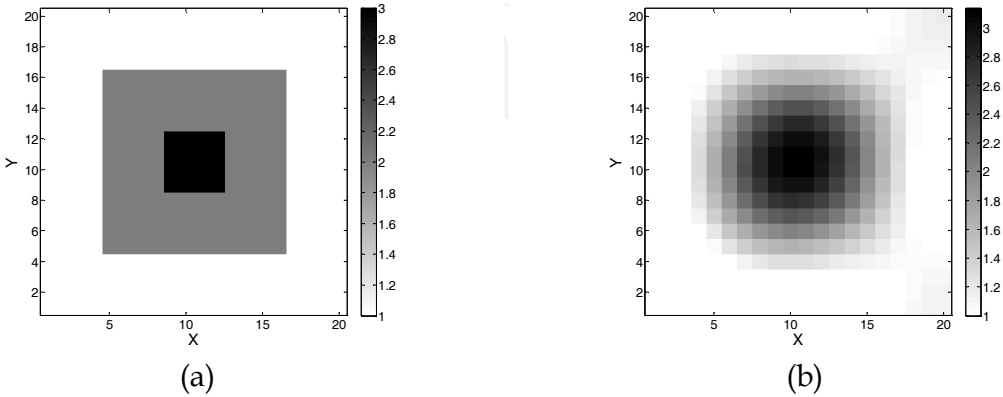


Fig. 8. Reconstruction of 2-D test case #1, (a) the cost function, (b) the reconstruction error

**Case study #2:** In this case, a lossy and inhomogeneous medium again with 20\*20 cells is considered. Therefore, we have two expansions for relative permittivity and conductivity profiles and in both expansions,  $N$  and  $M$  are chosen equal to 4, 5, 6 and 7. It is interesting to note that the number of direct optimization unknowns in this case is equal to 800 which is really a large optimization problem. The reconstructed profiles of permittivity and conductivity are shown in Figs. 9 and 10, respectively.



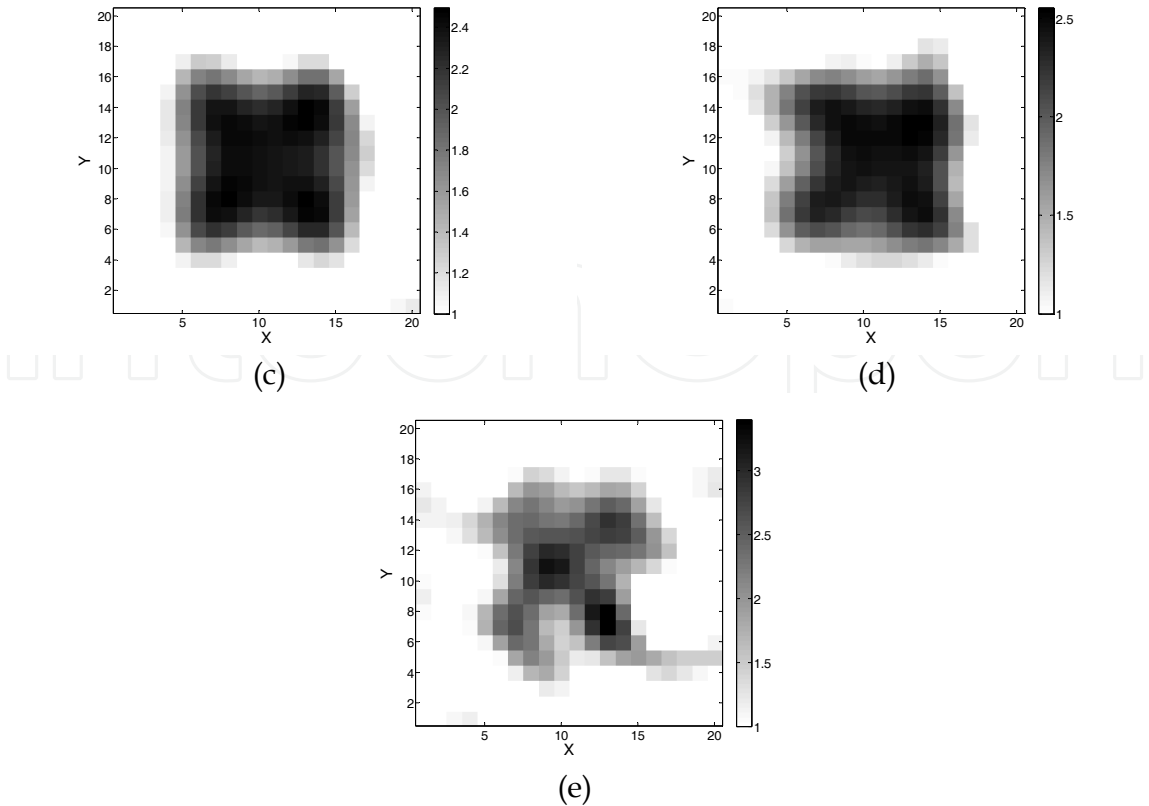
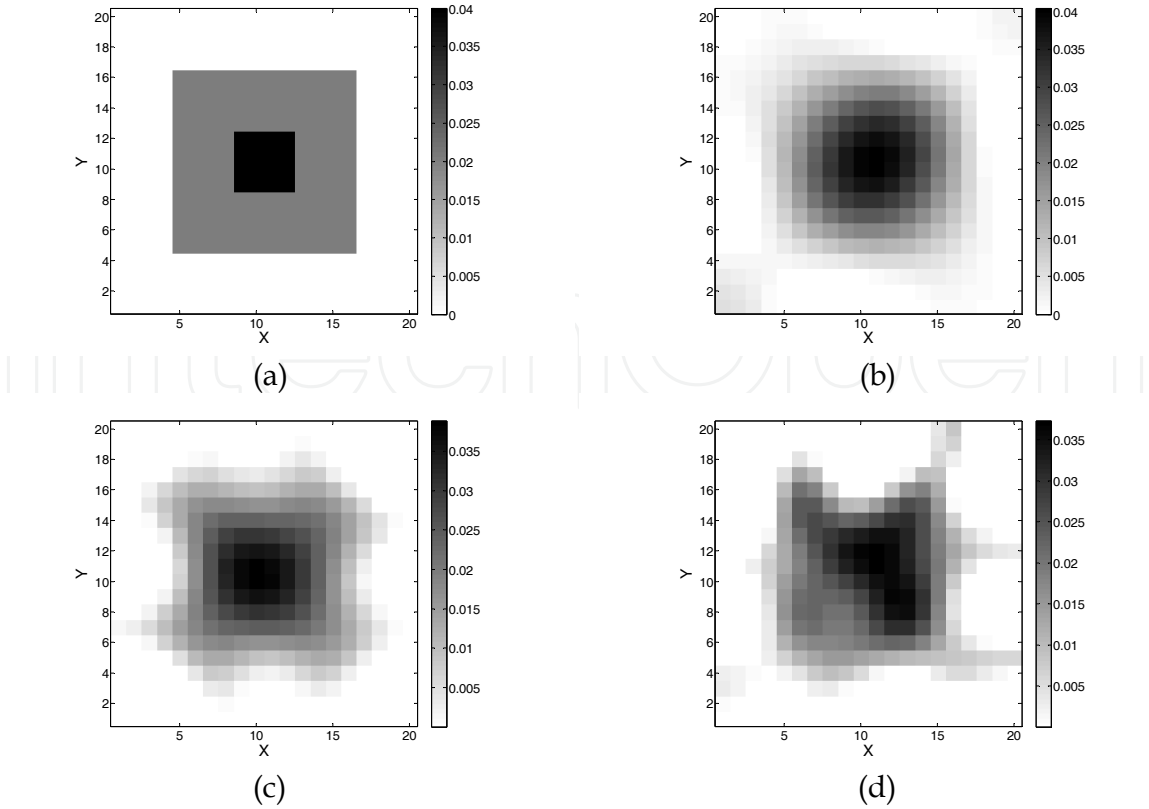


Fig. 9. Reconstruction of 2-D test case #2, (a) original permittivity profile, reconstructed permittivity profile with (b)  $N=M=4$ , (c)  $N=M=5$ , (d)  $N=M=6$  and (e)  $N=M=7$



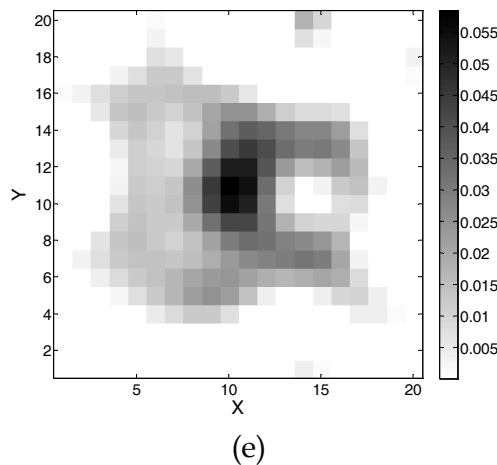
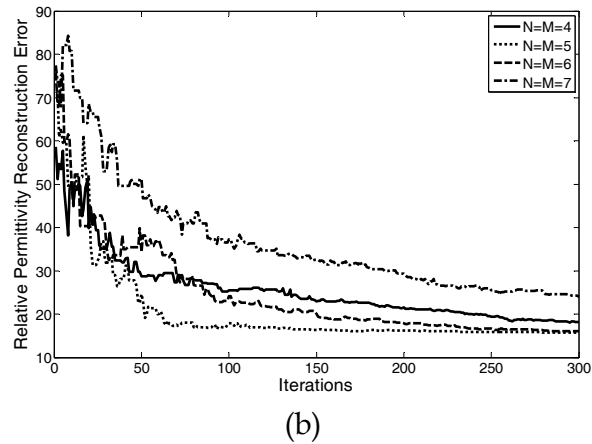
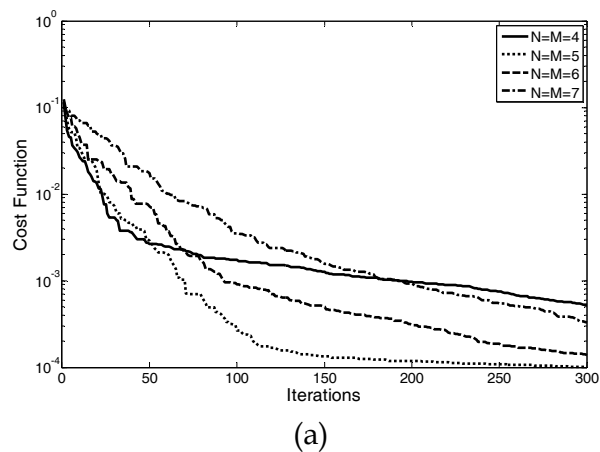


Fig. 10. Reconstruction of 2-D test case #2, (a) original conductivity profile, reconstructed conductivity profile with (b)  $N=M=4$ , (c)  $N=M=5$ , (d)  $N=M=6$  and (e)  $N=M=7$

The variations of cost function and reconstruction error are shown in Fig. 11.



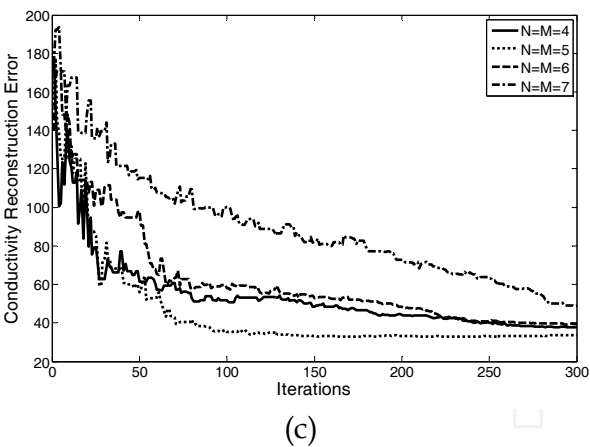


Fig. 11. Reconstruction of 2-D test case #2, (a) the cost function, (b) the permittivity reconstruction error and (c) the conductivity reconstruction error

The results of all 1-D and 2-D cases which are generally inhomogeneous and lossy or lossless media show that the proposed expansion method can tolerably reconstruct the unknown media with a considerable reduction in the amount of computations as compared to the conventional direct optimization of the unknowns.

5. Sensitivity Considerations

It is obvious that the performance of the expansion method directly depends on the number of expansion terms. Larger number of terms results in a more precise reconstruction at the expense of higher degree of ill-posedness. On the other hand, lower ones leads to a less accurate solution with higher probability of convergence of the inverse algorithm. Therefore, suitable selection of N has a notable impact on the convergence speed of the algorithm. The reconstructed profiles of two 1-D cases with larger values of N are shown in Figs. 12 and 13 for test case #1 and #2, respectively.

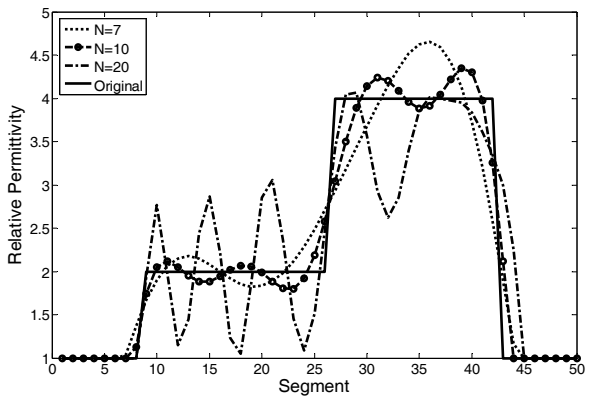


Fig. 12. Reconstruction of 1-D test case #1, the original profiles and reconstructed ones with N=7, 10 and 20

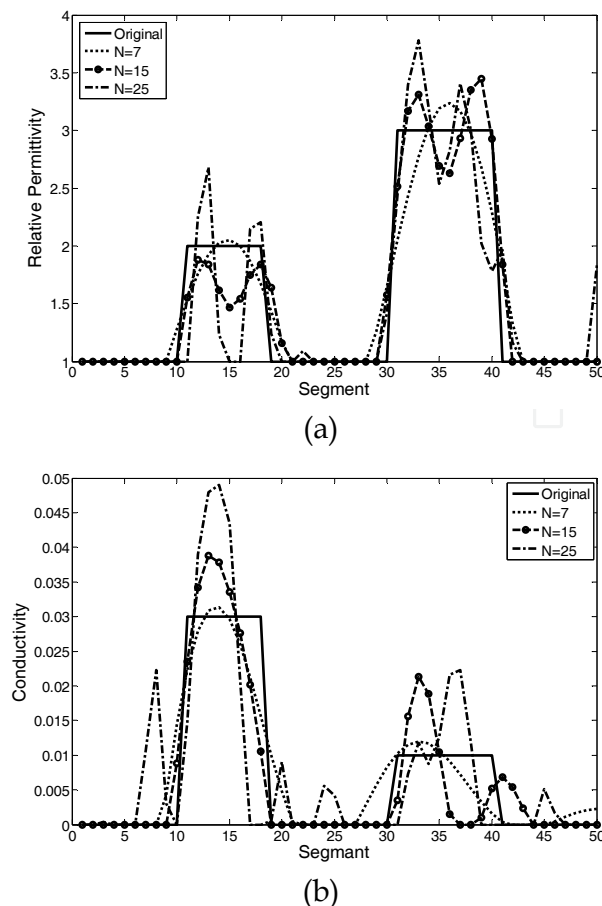


Fig. 13. Reconstruction of 1-D test case #2, the original profiles and reconstructed ones with  $N=7, 15$  and  $25$ , (a) permittivity profile and (b) conductivity profile

It is seen that increasing the number of expansion terms results oscillatory reconstruction because of the more ill-posedness of the problem.

We can come to similar conclusion for 2-D cases by comparing different parts of Figs. 7, 9 and 10.

Our experiences in studying various permittivity and conductivity profiles reconstruction show that choosing the number of expansion terms between 5 and 10 may be suitable for most of the reconstruction problems.

## 6. Conclusion

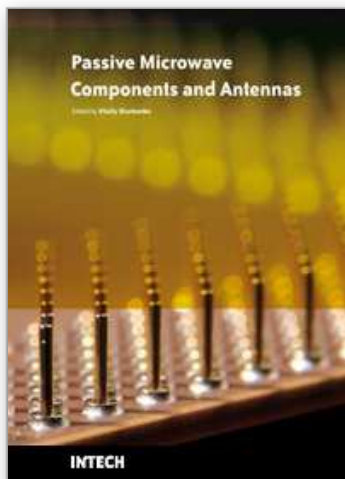
A computationally efficient method which is based on combination of the cosine Fourier series expansion, an EM solver and a global optimizer has been proposed for solving 1-D and 2-D inverse scattering problems. The mathematical formulations of the method have been derived completely and the algorithm has been examined for reconstruction of several inhomogeneous lossless and lossy cases. With a considerable reduction in the number of the unknowns and consequently the required number of populations and optimization iterations, along with no need to the regularization term, the relative permittivity and conductivity profiles have been reconstructed successfully. It has been shown by sensitivity



analysis that for obtaining well-posedness as well as accurate reconstruction simultaneously, the number of expansion terms must be chosen intelligently.

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## **Passive Microwave Components and Antennas**

Edited by Vitaliy Zhurbenko

ISBN 978-953-307-083-4

Hard cover, 556 pages

**Publisher** InTech

**Published online** 01, April, 2010

**Published in print edition** April, 2010

Modelling and computations in electromagnetics is a quite fast-growing research area. The recent interest in this field is caused by the increased demand for designing complex microwave components, modeling electromagnetic materials, and rapid increase in computational power for calculation of complex electromagnetic problems. The first part of this book is devoted to the advances in the analysis techniques such as method of moments, finite-difference time-domain method, boundary perturbation theory, Fourier analysis, mode-matching method, and analysis based on circuit theory. These techniques are considered with regard to several challenging technological applications such as those related to electrically large devices, scattering in layered structures, photonic crystals, and artificial materials. The second part of the book deals with waveguides, transmission lines and transitions. This includes microstrip lines (MSL), slot waveguides, substrate integrated waveguides (SIW), vertical transmission lines in multilayer media as well as MSL to SIW and MSL to slot line transitions.

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Abbas Semnani and Manoochehr Kamyab (2010). Solving Inverse Scattering Problems Using Truncated Cosine Fourier Series Expansion Method, *Passive Microwave Components and Antennas*, Vitaliy Zhurbenko (Ed.), ISBN: 978-953-307-083-4, InTech, Available from: <http://www.intechopen.com/books/passive-microwave-components-and-antennas/solving-inverse-scattering-problems-using-truncated-cosine-fourier-series-expansion-method1>

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